# Bioinformatic approaches to regulatory genomics and epigenomics

376-1347-00L | week 06

Pierre-Luc Germain



#### Plan

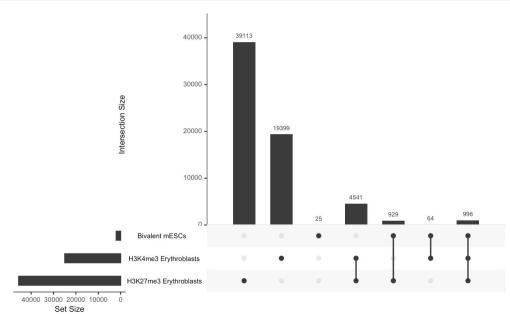
New packages to install (see slack)

Debriefing on last week's assignment

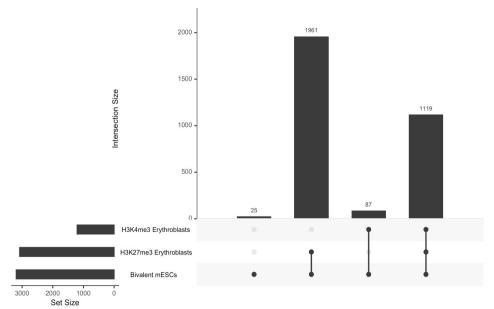
Overview of transcription factors and their binding specificity

DNA motifs and related analysis

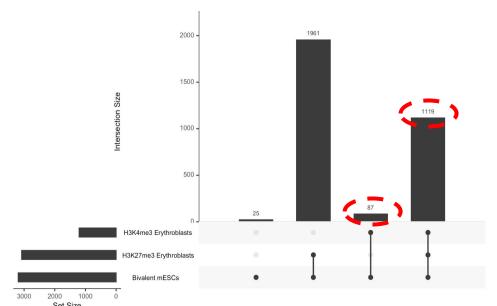
```
'``{r, without reference}
# without reference
peakList <- list(biValMe_2, H3K4me3_eb, H3K27me3_eb)
names(peakList) <- c("Bivalent mESCs", "H3K4me3 Erythroblasts", "H3K27me3 Erythroblasts")
regionUpset(peakList)
'``</pre>
```



```
'``{r, with reference}
# with reference
regionUpset(peakList, reference=peakList[[1]])
'``
```

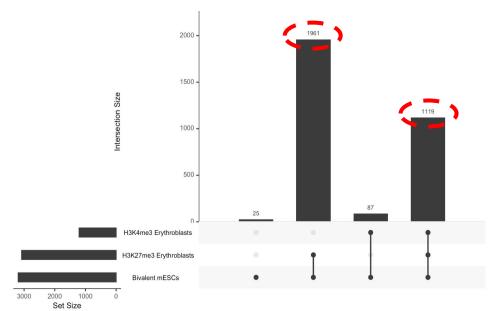


```
```{r, with reference}
# with reference
regionUpset(peakList, reference=peakList[[1]])
```
```



```
> sum(overlapsAny(biValMe_2, H3K4me3_eb))
[1] 1206
=87+1119
```

```
'``{r, with reference}
# with reference
regionUpset(peakList, reference=peakList[[1]])
'``
```



```
> sum(overlapsAny(biValMe_2, H3K27me3_eb))
[1] 3080
=1916+1119
```

When no reference is specified, one is created automatically by merging and *reducing* the regions (unless otherwise specified in the arguments):

| regions1                      |  |
|-------------------------------|--|
| regions2                      |  |
| reduce(c(regions1, regions2)) |  |

# Intersection & overlap: The example of bivalent domains

H3K4me3: H3K27me3: method one (overlapsAny): find the H3K4me3 peaks that overlap a H3K27me3 domain method two (intersect): find the regions that are covered by both H3K4me3 and H3K27me3

#### **Annotations:**

#### **ENCFF247GVM**

#### **Histone ChIP-seq in ES-Bruce4**

Mus musculus strain Bruce4 ES-Bruce4

Target: H3K4me3

Lab: Bing Ren, UCSD

**Project:** ENCODE

Reference Epigenome: ENCSR343RKY

candidate Cis-Regulatory Elements (cCREs): SCREEN

#### ENCFF326VMV

#### Histone ChIP-seq in smooth muscle cell

Homo sapiens smooth muscle cell originated from H9

Target: H3K4me3

Lab: Bradley Bernstein, Broad

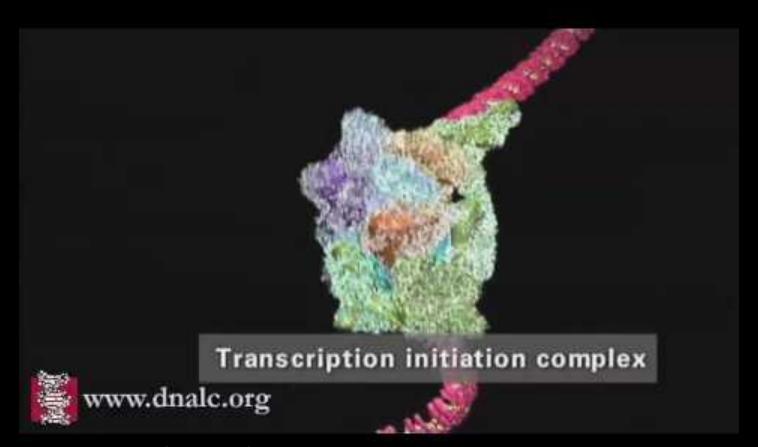
**Project:** ENCODE

**Reference Epigenome: ENCSR116JEF** 

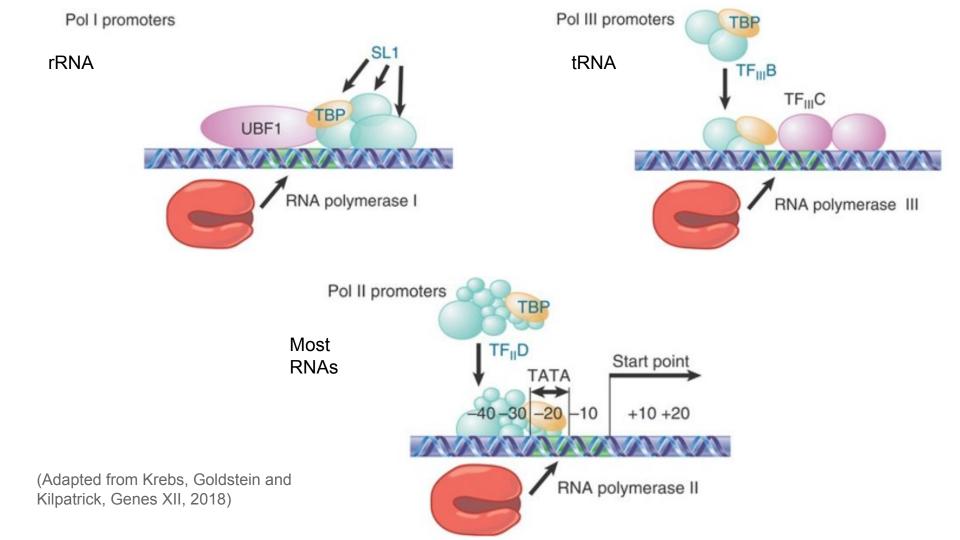
candidate Cis-Regulatory Elements (cCREs): SCREEN

```
regionUpset(peaks, nsets=length(peaks))
```

```
## Warning in .merge_two_Seqinfo_objects(x, y): Each of the 2 combined objects has sequence levels not in the oth
er:
## - in 'x': chr20, chr21, chr22
## - in 'y': chr4_GL456216_random, chrUn_GL456368, chrUn_GL456370, chrUn_GL456378, chrUn_JH584304, chrX_GL45623
3_random
## Make sure to always combine/compare objects based on the same reference
## genome (use suppressWarnings() to suppress this warning).
```

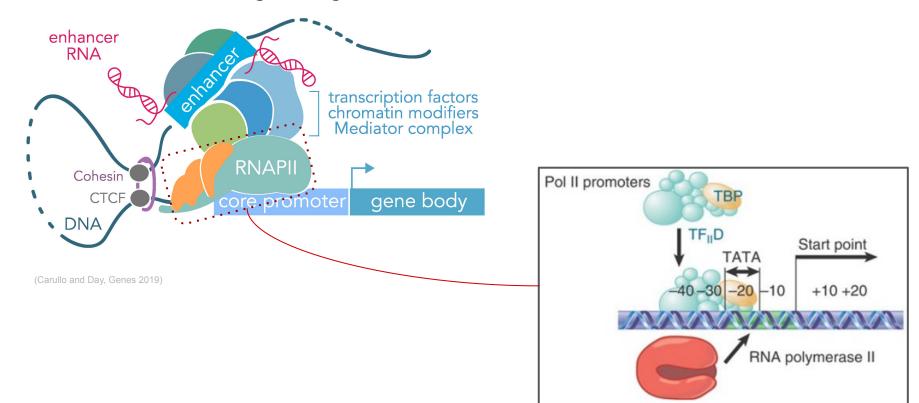


https://youtu.be/SMtWvDbfHLo



# Additional regulatory elements

#### Enhancer-driven gene regulation

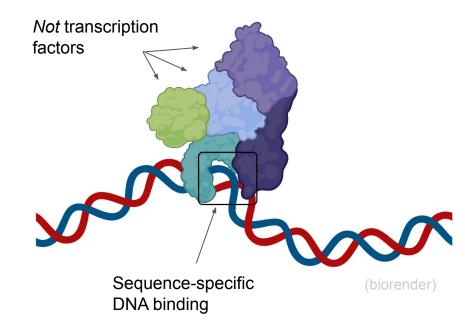


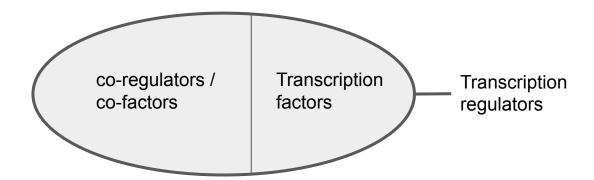
### What is a transcription factor?

#### Proteins capable of both:

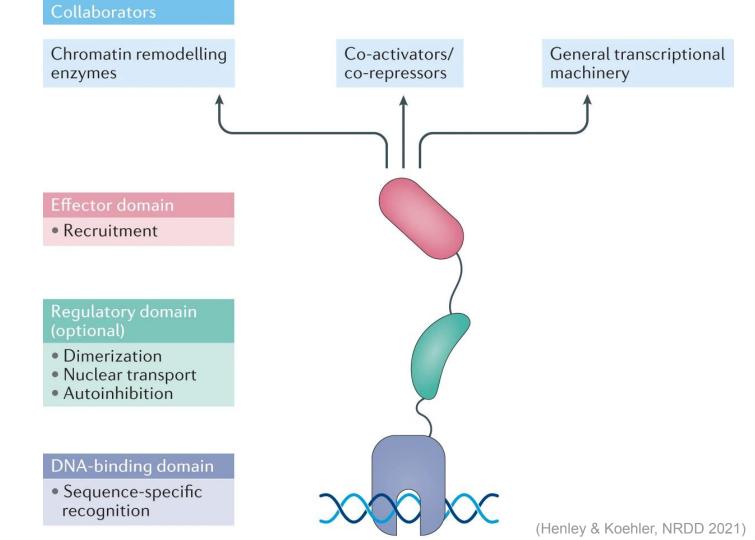
- 1) Binding DNA in a sequence-specific manner
- 2) Regulating transcription

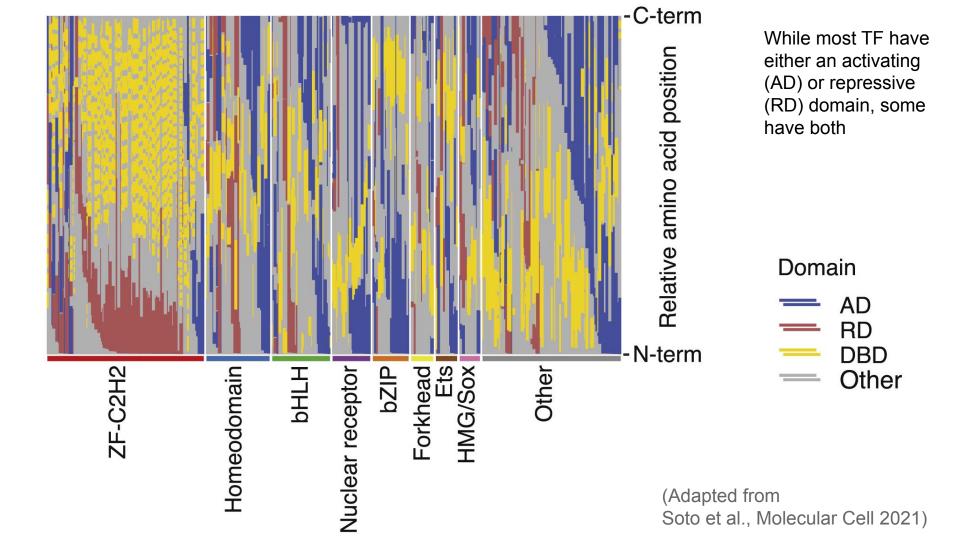
(Lambert et al., Cell 2018)





Anatomy of a transcription factor (TF)





Review (Cell 2018)

# The Human Transcription Factors

Samuel A. Lambert <sup>1, 9</sup>, Arttu Jolma <sup>2, 9</sup>, Laura F. Campitelli <sup>1, 9</sup>, Pratyush K. Das <sup>3</sup>, Yimeng Yin <sup>4</sup>, Mihai Albu <sup>2</sup>, Xiaoting Chen <sup>5</sup>, Jussi Taipale <sup>3, 4, 6</sup>  $\bowtie$   $\bowtie$ , Timothy R. Hughes <sup>1, 2</sup>  $\bowtie$   $\bowtie$ , Matthew T. Weirauch <sup>5, 7, 8</sup>  $\bowtie$ 

Proteins capable of both:

- 1) Binding DNA in a sequence-specific manner
- 2) Regulating transcription

According to their census, humans have 1570 transcription factors

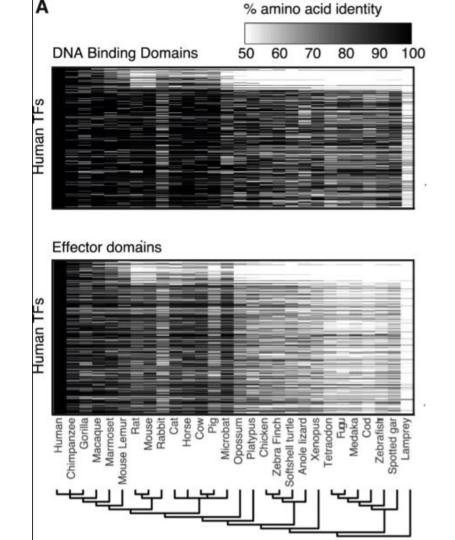
78 TFs with Multiple DBDs

713 TFs with C2H2 ZF arrays

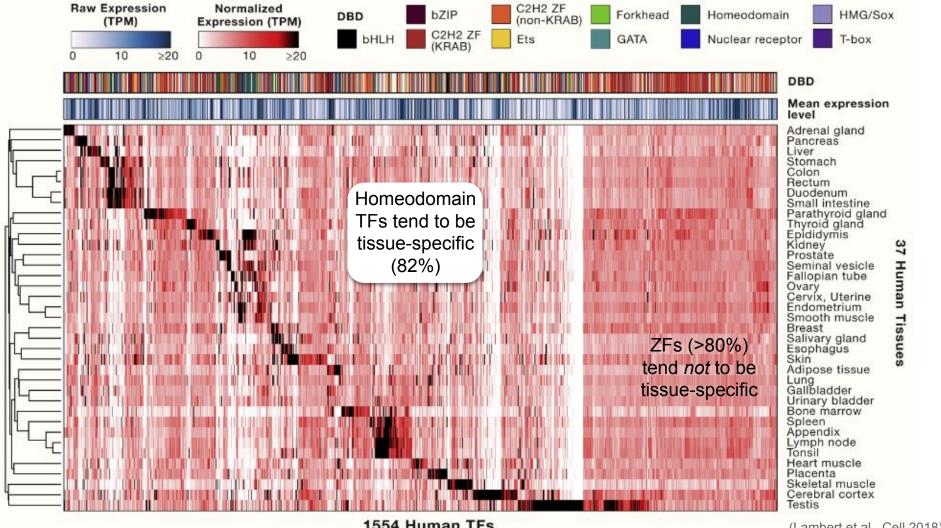
779 TFs with a single DBD

# Transcription factors are highly conserved

DNA binding domains show much higher conservation than effector domains



(Soto et al., Molecular Cell 2021)



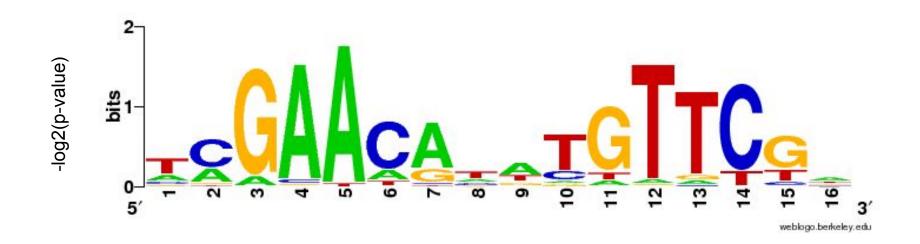
1554 Human TFs

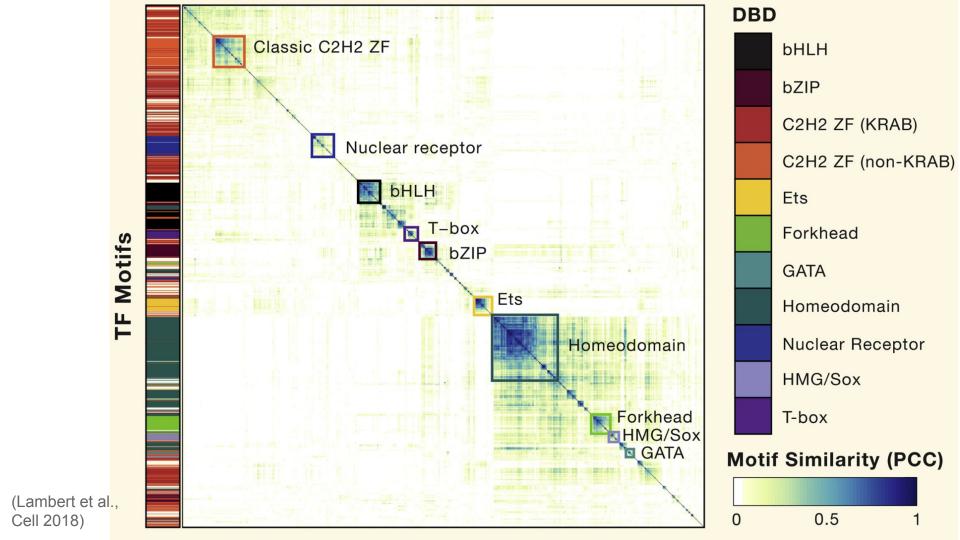
(Lambert et al., Cell 2018)

## Sequence-specificity

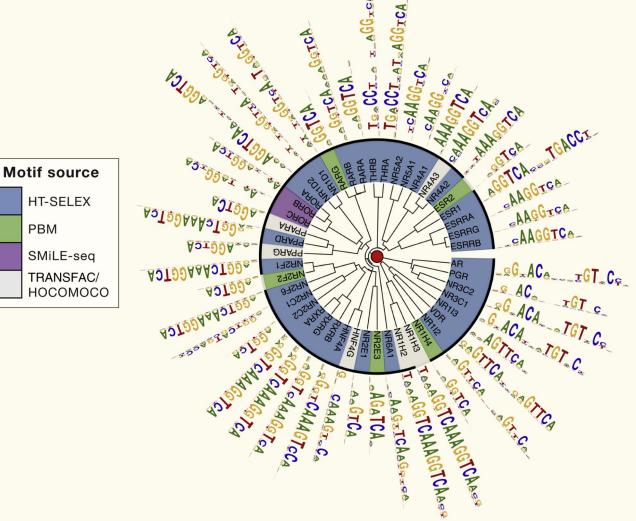
E.g. The LexA bacterial TF recognizes the consensus sequence

5'-GAACAnnTGTTC-3'

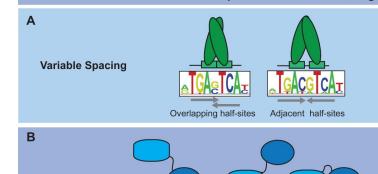




# An example of TF motif degeneracy: Nuclear hormone receptors



# Variations in DNA binding specificity



POU<sub>HD</sub> site

variable-length spacers (82); motifs from (73,74)

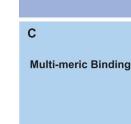
of its two DNA-binding domains (91,92);

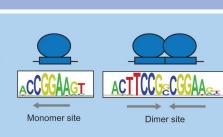
Gcn4 dimers can bind to bipartite

sites with half-sites separated by

Oct-1 can bind to different DNA sites using different arrangements

motifs from (24)





POU<sub>s</sub> site

POU site

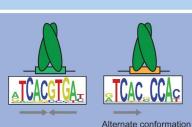
**Multiple Modes of DNA Binding** 

Elk1 can bind both as a monomer or as a dimer (95)

D

Alternate Structural Conformations

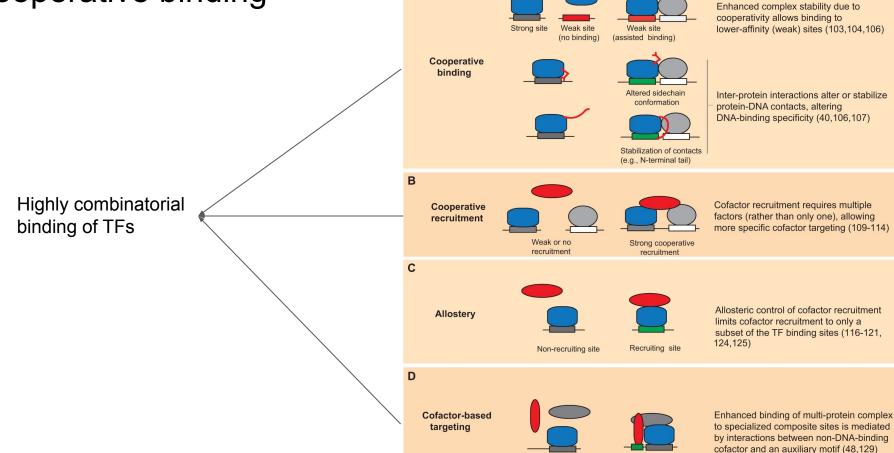
**Multiple DBDs** 



SREBP can bind to different DNA sites by adopting alternate structural conformations (96,97); motifs from (44)

(Siggers and Gordân, NAR 2014)

# Cooperative binding



A

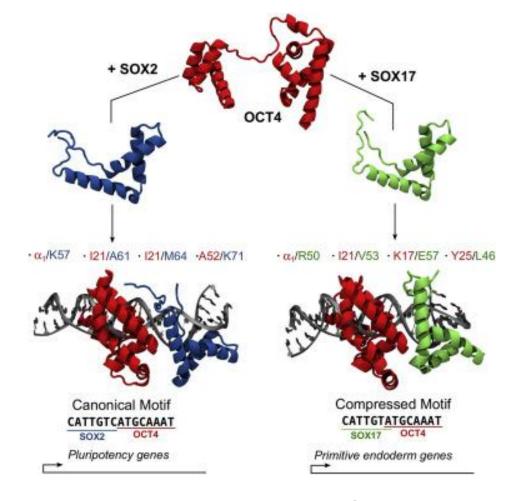
**Multi-Protein Recognition Codes** 

Enhanced binding to composite site

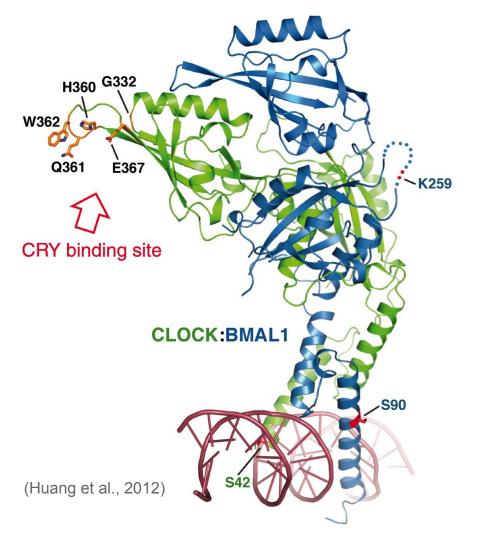
(Siggers and Gordân, NAR 2014)

# Two examples of Cooperative binding

OCT4 (POU5f1) binding upon differentiation

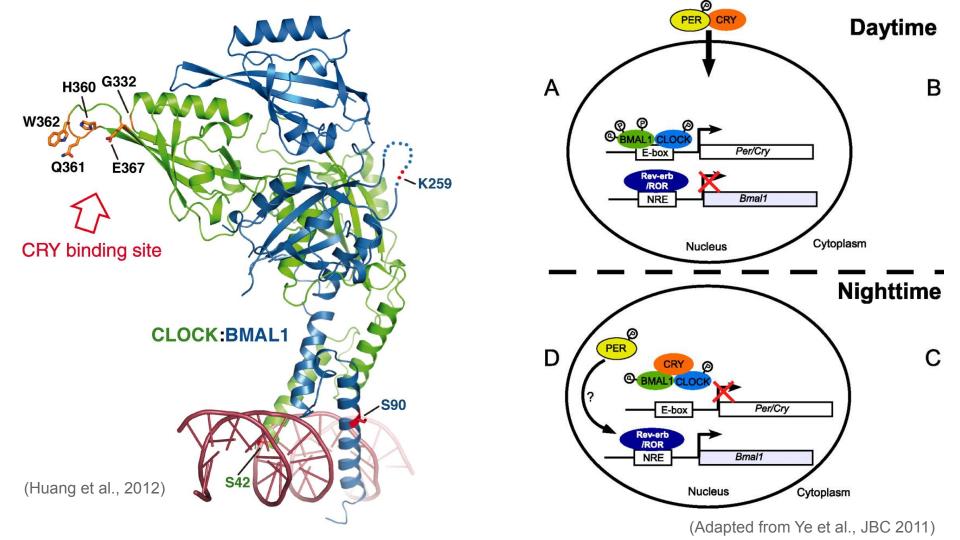


(Merino et al., Structure 2014)



# Clock-Bmal-Cry during circadian rythm





## Motif analysis

- Motif discovery aims at finding new motifs that are enriched in a set of sequences (e.g. peaks)
  versus a background
  - Example method: meme (Meme suite)
  - Bioconductor method: rGADEM package (see also the memes R package)
- Motif enrichment analysis aims at finding known motifs that are enriched in a set of sequences (e.g. peaks) versus a background
  - Example method: AME (Meme suite)
  - Bioconductor method: PWMEnrich package
- Motif scanning aims at finding the occurrences of known motifs in a set of sequences (methodologically fairly simple – which method doesn't matter much)
  - Example method: fimo (Meme suite)
  - Bioconductor method: motifmatchr (see also TFBSTools package)

# Genetic variation at TF binding sites

- Genetic variation at TF binding sites can affect the binding of the protein, and hence impact development and health
- Nevertheless, while most coding sequences show evidence of evolutionary constraint (e.g. purifying selection), only a small fraction of TF binding sites (11.6% of footprints) show evidence of constraint – the vast majority appears to be evolving neutrally

(Vierstra et al., Nature 2020)

This suggests a degree of (at least partial) redundancy between regulatory elements

# Assignment

- Choose a transcription factor, e.g. CREB1, REST, GATA5, EGR1, GCR (or any of your choice that has a motif and available ChIPseq data)
- Download the (e.g. Mouse) peaks for that factor (whatever cell type)
- Identify the instances of the factor's motif
- Answer the following questions:
  - Of all the peaks, what proportion contains a motif for the factor?
    - Expected form of an answer: of the XX peaks, XX (XX%) contain a motif
  - Of all instances of that motif in the genome (or in one chromosome), what proportion is bound by the factor (i.e. has a peak)?
    - Expected form of an answer: of the XX motif instances, XX (XX%) overlap a peak

Don't forget to render your markdown and push it as assignment.html!